

A Generative Constraint Model for Optimizing Software Deployment

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Motivation

- More and more functions in today's cars involve electronics and software
- 80-90 percent of the new innovative features are realized by distributed embedded systems
- Today's upper class cars contain up to 80 ECUs (Electronic Control Units)
- Even highly safety critical mechanical and hydraulic control systems will be replaced by electronic components



Software Deployment

- Today's embedded automotive software is highly distributed
- The automotive industry devotes increasing efforts to develop tools for automated software deployment.
- The underlying foundations comprise techniques like genetic algorithms and clustering techniques.
- Measuring the quality of the deployment process: Metrics



Metrics

Measure the quality of the deployment process

- Bus-load: The system's ECUs are connected by means of data-pipes or buses.
 - It is a critical metric of the system. It's value should be as low as possible. Otherwise: bottlenecks.
- Resources: The number and type of ECUs required to implement the functionalities.
- Costs: The required financial effort for a configuration.



Function Blocks

- Function block: a specific task from a software functionality of the system.
- Example: The Signaling Function
 - 1 Hit the commutator (Change State)
 - 2 Select the targeted component (Observer)
 - 3 Start blinking
- The function blocks have to exchange data ⇒ The network of function blocks.
- Each branch from the network has a weight, the communication frequency (CF).





Clusters

- Cluster = a set of function blocks grouped together.
- To each cluster corresponds an ECU for execution.
- Clusters should contain optimal function distribution.
- Apply quality criteria in order to create the best clustering schema.



Constraint Satisfaction Problem

A Constraint Satisfaction Problem (CSP) is a tuple (V, D, CO).

1. { var_1 = (
$$x_0 < y_0$$
);

2.
$$\min_{-1} = x_{-0};$$

3. min_2 = y_0; }

Variables: $V = \{var_1, x_0, y_0, min_1, min_2\}$ Domains: $D = \{D(x) = N | x \in V\}$ Constraints:

$$CO = \left\{ \begin{array}{c} var_{-}1 = (x_{-}0 < y_{-}0), \\ min_{-}1 = x_{-}0, min_{-}2 = y_{-}0 \end{array} \right\}$$

 State of the art constraint solvers are available for solving CSPs.



CSP Partitioning

- The partitioning problem: How to group the function blocks into clusters?
- Transform the problem of cluster partitioning into a CSP:
 - Use the available resources
 - Use Quality Functions
 - Use the cost limitations



Resource Parametrization

• Conversion to $CSP \Rightarrow$ defining the tuple (*V*, *D*, *CO*)

- Function blocks, $F = \{f_1...f_t\}$.
- The communication frequency sets $CF = \{CF_1...CF_t\}, CF_i = \{cf_{i1}, ...cf_{it}\}.$
- Available set of ECUs ECU = {ECU₁...ECU_k} characterized by:
 - Memory: $MEM = \{mem_1...mem_k\}$
 - Processing power: $PROC = \{proc_1...proc_k\}$



Resource Constraints

- 1 The overall memory consumption of the function blocks is smaller or equal to the available memory:
- 2 Maximal function block memory constraint
- Each ECU must have enough resources to accommodate the assigned cluster
- A function block is deployed on a single ECU only
- 5 Any function *deploy* that distributes all functional blocks f_i on *max* ECUs is a solution.



Timing Constraints

- We assume that:
 - There is only a single path from the source x to the sink y
 - That there are no loops in the network.
- Then we must have:

1
$$t_{xy} \leq \sum_{i} t_{wc}(f_i) + t_{wc}(B_i) + t_{wc}(G_i).$$

2 SF:
$$TSK = \{t_1, ..., t_k\} \iff \sum_{i} time(t_i) \le T_{SF}$$

3 t_{ECUi}: the time that a ECU needs to execute a task t_i then

 $t_{FCUi} < time(t_i).$

For every timing requirement $Req_{(a,b)}$ we instantiate these constraints.



Quality Constraints

- The most important factor when we partition the function blocks into clusters.
- We use predefined quality functions.
- Constraint System building: output values that no cluster is allowed to exceed.
- Each quality function receives as input parameter the CF set.



Quality Functions

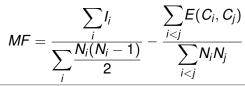
The External-Internal Ratio: ratio between the external and the internal costs must be as low as possible.

$$\forall C_i, i \in [1, c] \frac{E_i}{I_i}$$

The Davies Bouldin Criteria: shows a good partitioning when the factor is as low as possible.

$$\mathcal{DB} = rac{1}{c} \sum_{i=1}^{c} max_{j
eq i} \left[rac{diam(C_i) + diam(C_j)}{d(C_i, C_j)}
ight]$$

The Modularization Factor: indicator of a compact clustering. Should be as high as possible.





The SILHOUETTE factor: correctness of the distribution of a function f_i within a cluster C_i with respect to a neighbor node C_j. A good factor is close to 1 :

$$Sh(f_i) = \frac{d(f_i,C_j) - d(f_i,C_i)}{\max(d(f_i,C_j),d(f_i,C_i))}$$

The Cluster Load Deviation: as low as possible. In a good partitioning similar number of function blocks in all clusters:

$$CLD = \sqrt{\frac{1}{c-1}\sum_{i=1}^{c}(N_i - \bar{N})^2}$$



Cost Constraints

- Built based on the system's cost criteria.
- Each ECU has a price and a performance description associated to it.
- The Price Constrain: Distribute the function block set such that the total cost of the ECUs is smaller than a given price.
- The Bus Load Constraint: At any given moment the bus load should not exceed a critical imposed value.

 $CSP = RCS \cup QCS \cup CCS \cup TCS.$



Generative Deployment of Clusters

- Different types of ECUs on which we must assign the clusters.
- The requirements of this cluster are generated by means of the generative constraints.
- Choose from the set of ECUs the cheapest ECU and try to deploy the cluster.
- If after replacing the meta variables with the concrete variables, an inconsistency happens ⇒

Choose the next best ECU by using the nogoods provided by the consistency checker



Conclusions

- (CSP) representations are successfully used in diverse areas from configuration and reconfiguration of large systems.
- We outline a novel modeling approach that allows deployment of embedded automotive software.
- The Model-based approach allows for automatic deployment of software functions in a resource-constrained software system.
- Computation of a valid solution satisfying the outlined criteria by relying on standard CSP solvers.
- We do not generate all the cluster combinations but rely on the the first n, solutions that the CSP solver comes up with.



Thank You!

Q & A

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